



Two segmentation methods

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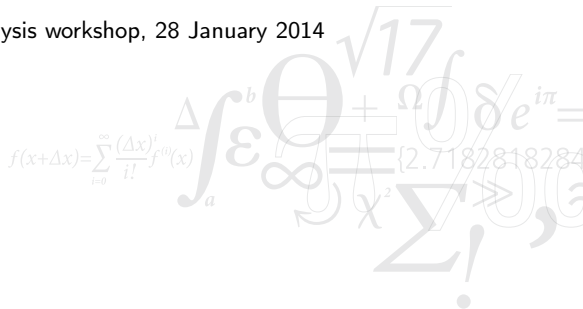
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Two segmentation methods

Vedrana Andersen Dahl

DTU Compute

Image reconstruction and analysis workshop, 28 January 2014

$$f(x+\Delta x) = \sum_{i=0}^{\infty} \frac{(\Delta x)^i}{i!} f^{(i)}(x)$$


Presentation overview

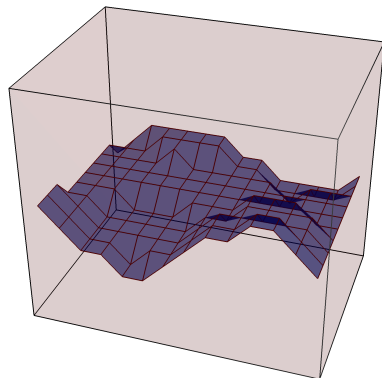
- ▶ RoundCut: volumetric segmentation using mesh-based optimal graph search
 - ▶ (with Anders Bjorholm Dahl and Rasmus Larsen)
- ▶ Multiphase image segmentation using the deformable simplicial complex method
 - ▶ (with Asger Nyman Christiansen and Jakob Andreas Bærentzen)
- ▶ ... combining the two?

$$f(x+\Delta x) = \sum_{i=0}^{\infty} \frac{(\Delta x)^i}{i!} f^{(i)}(x)$$

Optimal surface search

Problem

- ▶ 3D volume
- ▶ terrain-like surface
- ▶ smoothness constraint
- ▶ optimality



Solution^{1,2}

- ▶ graph based
- ▶ allowing for a number of useful extensions (including RoundCut)

¹Xiaodong Wu and Danny Z Chen. Optimal net surface problems with applications. In *Automata, Languages and Programming*, pages 1029–1042. Springer, 2002

²Kang Li, Xiaodong Wu, Danny Z Chen, and Milan Sonka. Optimal surface segmentation in volumetric images—a graph-theoretic approach. *TPAMI*, 28(1):119–134, 2006

Algorithm fundamentals

Optimal net surface problem

- ▶ 3D volume, cost

$$c(x, y, z)$$

- ▶ terrain-like surface

$$z = S(x, y)$$

- ▶ smoothness constraint

$$|S(x, y) - S(x - 1, y)| \leq \Delta_x$$

$$|S(x, y) - S(x, y - 1)| \leq \Delta_y$$

- ▶ optimality

$$\min_S \sum_{x,y} c(x, y, S(x, y))$$

Minimum closed set problem (in a vertex weighted graph)



Minimum s - t cut problem (in an edge weighted graph)³



³Yuri Boykov and Vladimir Kolmogorov. An experimental comparison of min-cut/max-flow algorithms for energy minimization in vision.

Algorithm fundamentals

- ▶ 3D volume, cost

$$c(x, y, z)$$

- ▶ terrain-like surface:

base arcs, inter-column arcs

$$z = S(x, y)$$

- ▶ smoothness constraint:

intra-column arcs

$$|S(x, y) - S(x - 1, y)| \leq \Delta_x$$

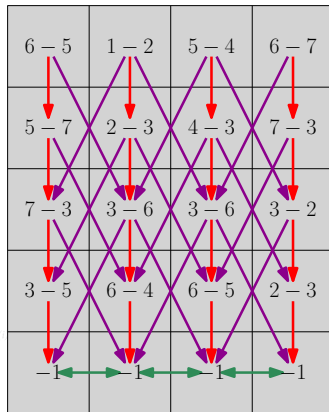
$$|S(x, y) - S(x, y - 1)| \leq \Delta_y$$

$$\Delta_x = 2$$

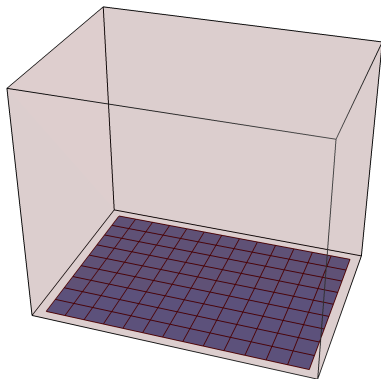
- ▶ optimality:

weights

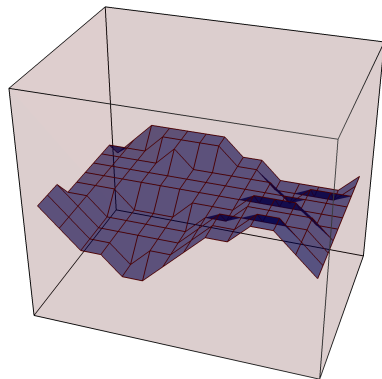
$$\min_S \sum_{x,y} c(x, y, S(x, y))$$



Terrain-like surfaces



Base surface

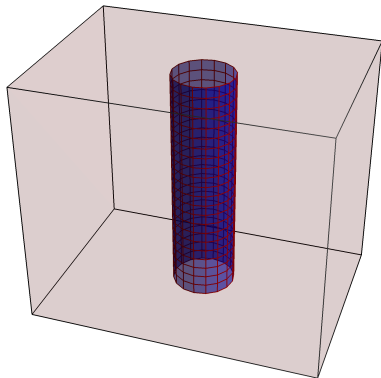


Realization

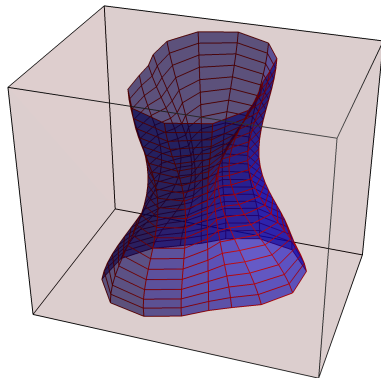
$$f(x) = \sum_{i=0}^{\infty} \frac{f_i(x)}{2^i}$$

$$i\pi = 3284$$

Tubular surfaces



Base surface

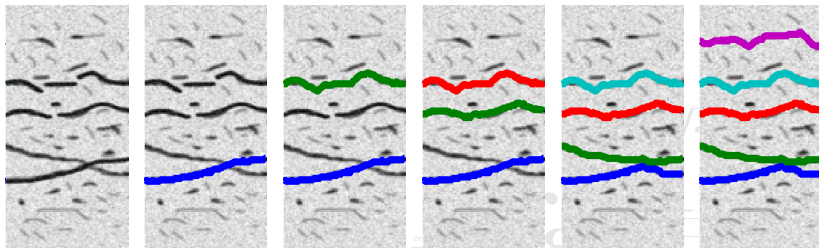


Realization

$$)=\sum_{i=0}^{\infty} \binom{r}{i}$$

$$i\pi = 3284$$

Multiple interrelated surfaces



Extensions

- ▶ In-region cost ⁴
- ▶ Varying smoothness constraints ⁵
- ▶ Shape and context priors ⁶

⁴Mona Haeker, Xiaodong Wu, Michael Abràmoff, Randy Kardon, and Milan Sonka.

Incorporation of regional information in optimal 3-d graph search with application for intraretinal layer segmentation of optical coherence tomography images.

In *Information Processing in Medical Imaging*, pages 607–618. Springer, 2007

⁵Mona Haeker, Michael D Abràmoff, Xiaodong Wu, Randy Kardon, and Milan Sonka. Use of varying constraints in optimal 3-d graph search for segmentation of macular optical coherence tomography images.

In *MICCAI*, pages 244–251. Springer, 2007

⁶Qi Song, Junjie Bai, M Garvin, Milan Sonka, J Buatti, and Xiaodong Wu. Optimal multiple surface segmentation with shape and context priors.

TMI, 32(2):376–386, 2013

Segmentation

○

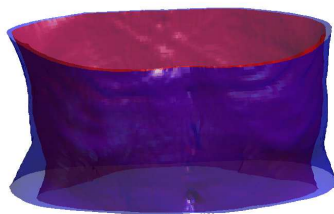
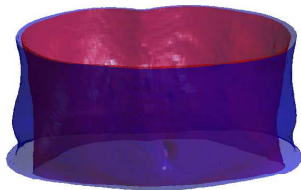
RoundCut

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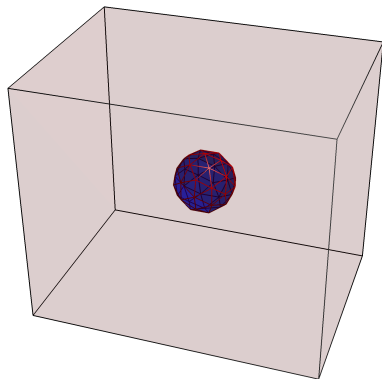
DSC

○○○○○

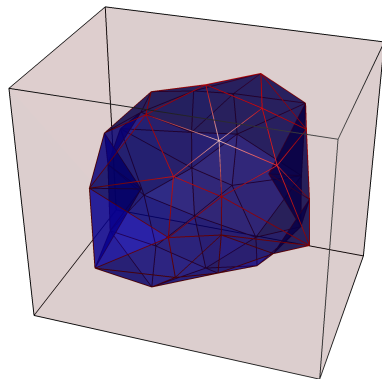
AbFat example



RoundCut



Base surface



Realization

$$)=\sum_{i=0}^{\infty}(-1)^i$$

$$i\pi =$$

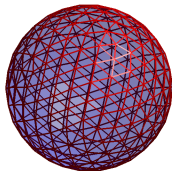
$$3284$$

$$3$$

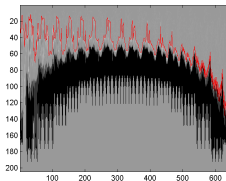
RoundCut basics

Approach related to:

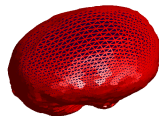
- ▶ NuggetCut⁷
- ▶ Radial ray based segmentation⁸



Initialization



Unwrapped image



Solution

⁷Jan Egger, Miriam HA Bauer, Daniela Kuhnt, Barbara Carl, Christoph Kappus, Bernd

Freisleben, and Christopher Nimsy. Nugget-cut: a segmentation scheme for spherically-and elliptically-shaped 3d objects.

In *DAGM Pattern Recognition*, pages 373–382. Springer, 2010

⁸Sebastian Steger, Nazli Bozoglu, Arjan Kuijper, and Stefan Wesarg. Application of radial ray based segmentation to cervical lymph nodes in ct images.

TMI, 32(5), 2013

Segmentation



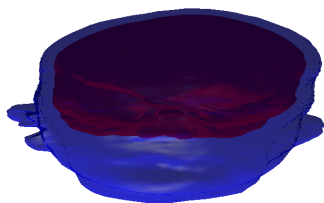
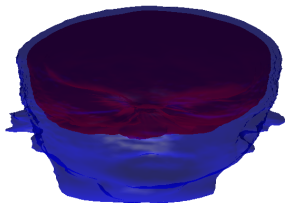
RoundCut

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DSC

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ICV results



RoundCut

- ▶ Plus: generality, arbitrary topology and geometry
- ▶ Minus: loosing matrix representation
- ▶ Plus: hierarchical or iterative approach

$$f(x+\Delta x) = \sum_{i=0}^{\infty} \frac{(\Delta x)^i}{i!} f^{(i)}(x)$$

$$\int_a^b \varepsilon \Theta + \Omega \int \delta e^{i\pi} = \{2.7182818284\}$$

$$\chi^2 \sum_i !$$

Iterative RoundCut: beyond the star domain

$$f(x+\Delta x) = \sum_{i=0}^{\infty} \frac{(\Delta x)^i}{i!} f^{(i)}(x)$$

$$\Delta \int_a^b \varepsilon \Theta + \Omega \int \delta e^{i\pi} = \{2.7182818284\}$$

$$\chi^2 \sum_i !$$

Iterative RoundCut: beyond the star domain

$$f(x+\Delta x) = \sum_{i=0}^{\infty} \frac{(\Delta x)^i}{i!} f^{(i)}(x)$$

$$\Delta \int_a^b \varepsilon \Theta + \Omega \int \delta e^{i\pi} = \{2.7182818284\}$$

$$\chi^2 \sum_i !$$

Many challenges and a few solutions

- ▶ Folding when normal sampling (gradient vector flow⁹? flow lines¹⁰? mesh smoothing?)
- ▶ Applicability of the smoothness constraint
 - ▶ Irregular mesh (adaptive sampling rate? mesh optimization?)
 - ▶ Accumulation of differences (mesh smoothing?)
- ▶ Handling interrelated surfaces (maintaining correspondence?)
- ▶ Explicit mesh processing
 - ▶ Mesh subdivision (Loop and butterfly)
 - ▶ Mesh smoothing (implicit and explicit Laplacian smoothing)

⁹Christian Bauer, Shanhui Sun, and Reinhard Beichel. Avoiding mesh folding in 3d optimal surface segmentation.

In *Advances in Visual Computing*, pages 214–223. Springer, 2011

¹⁰Jens Petersen, Mads Nielsen, Pechin Lo, Zaigham Saghir, Asger Dirksen, and Marleen

De Bruijne. Optimal graph based segmentation using flow lines with application to airway wall segmentation.

In *Information Processing in Medical Imaging*, pages 49–60. Springer, 2011

Micro-CT scan



Segmentation

○

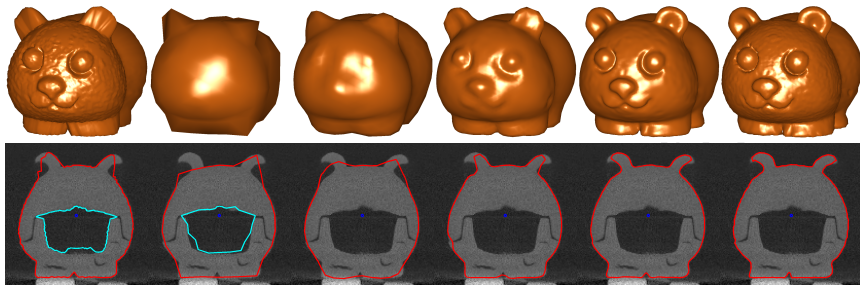
RoundCut

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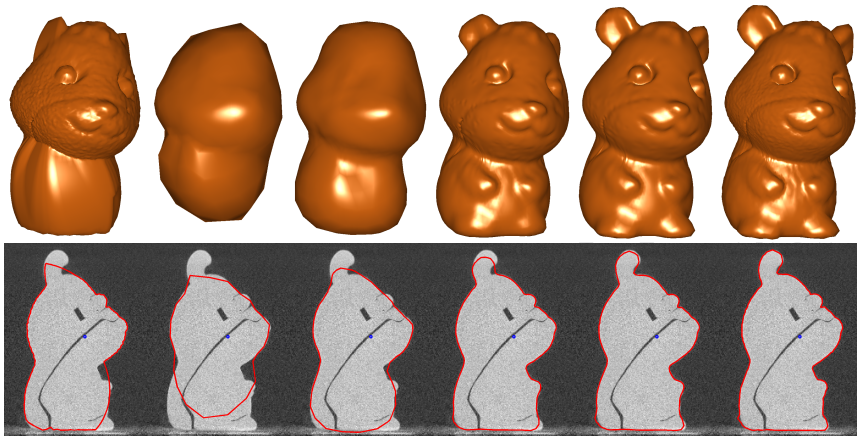
DSC

○○○○○

Micro-CT scan



Micro-CT scan



Structural and Multidisciplinary Optimization, pages 1–13, 2013

Deformable models for image segmentation

- ▶ Curves or surfaces that can move under:
 - ▶ internal forces (smoothness or other prior info)
 - ▶ external forces (image or volume data)
- ▶ Depending of the representation of the interface:
 - ▶ parametric deformable models, explicit interface (Lagrangian)
 - ▶ geometric deformable models, implicit interface (Eulerian)

$$f(x+\Delta x) = \sum_{i=0}^{\infty} \frac{(\Delta x)^i}{i!} f^{(i)}(x)$$

$$\int_a^b \varepsilon \Theta = \Omega \int \delta e^{i\pi} = \{2.7182818284\}$$

$$\chi^2$$

$$\Sigma!$$

ImSeg 2DDSC application

Features:

- ▶ Explicit curve tracking (advantages: boundary length, volume and size estimation)
- ▶ Adaptive topology (or not!)
- ▶ Multi phase
- ▶ Currently: region based external force ^{15, 16}

$$f(x+\Delta x) = \sum_{i=0}^{\infty} \frac{(\Delta x)^i}{i!} f^{(i)}(x)$$

¹⁵Tony F Chan and Luminita A Vese. Active contours without edges. *IEEE Transactions on Image Processing*, 10(2):266–277, 2001

¹⁶Andy Tsai, Anthony Yezzi Jr, and Alan S Willsky. Curve evolution implementation of the Mumford-Shah functional for image segmentation, denoising, interpolation, and magnification. *IEEE Transactions on Image Processing*, 10(8):1169–1186, 2001

DSC segmentation

- DSC interface tracking

$$f(x+\Delta x) = \sum_{i=0}^{\infty} \frac{(\Delta x)^i}{i!} f^{(i)}(x)$$

$$\Delta \int_a^b \varepsilon \Theta + \Omega \int \delta e^{i\pi} = \{2.7182818284\}$$

$$\chi^2 \sum_i \gg \infty$$

$$\sqrt{17} f$$

$$\Sigma_i !$$

DSC segmentation

- Adaptive topology

$$f(x+\Delta x) = \sum_{i=0}^{\infty} \frac{(\Delta x)^i}{i!} f^{(i)}(x)$$

$$\Delta \int_a^b \varepsilon \Theta + \Omega \int \delta e^{i\pi} = \{2.7182818284\}$$

$$\chi^2 \sum_i \gg$$

DSC segmentation

- Level of detail

$$f(x+\Delta x) = \sum_{i=0}^{\infty} \frac{(\Delta x)^i}{i!} f^{(i)}(x)$$
$$\Delta \int_a^b \varepsilon \Theta + \Omega \int \delta e^{i\pi} = \{2.7182818284\}$$
$$\chi^2 \sum_i !$$

DSC segmentation

- Multi phase support

$$f(x+\Delta x) = \sum_{i=0}^{\infty} \frac{(\Delta x)^i}{i!} f^{(i)}(x)$$

$$\Delta \int_a^b \varepsilon \Theta + \Omega \int \delta e^{i\pi} = \{2.7182818284\}$$

$$\chi^2 \sum_i !$$

DSC segmentation

- Level of detail

$$f(x+\Delta x) = \sum_{i=0}^{\infty} \frac{(\Delta x)^i}{i!} f^{(i)}(x)$$
$$\Delta \int_a^b \varepsilon \Theta + \Omega \int \delta e^{i\pi} = \{2.7182818284\}$$
$$\chi^2 \sum_i !$$

Challenges

- ▶ Two-phase region force

$$\mathbf{F}_{\text{ext}}(\mathbf{X}) = (m_{\text{out}} - m_{\text{in}})(I - m_{\text{out}} + I - m_{\text{in}})\mathbf{N}$$

and multiphase region force

$$\mathbf{F}_{\text{ext}}(\mathbf{X}_{ij}) = \frac{I - m_i + I - m_j}{m_i - m_j} \mathbf{N}_{ij}$$

- ▶ Forces on tripple junctions and other crossings
- ▶ Criterion for explicit topology change
- ▶ Regularization vs. DSC optimization
- ▶ Formalization in variational method framework: minimizing Mumford-Shah functional.
- ▶ 3D!

Thank you!

$$f(x+\Delta x) = \sum_{i=0}^{\infty} \frac{(\Delta x)^i}{i!} f^{(i)}(x)$$

$$\Delta \int_a^b \varepsilon \Theta + \Omega \int \delta e^{i\pi} = \{2.7182818284\}$$

$$\chi^2 \sum_i !$$